

IN THE UNITED STATES PATENT AND TRADEMARK OFFICE
APPLICATION FOR LETTERS PATENT

**Derivation and Quantization of Robust Non-
Local Characteristics for Blind
Watermarking**

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ATTORNEY's DOCKET NO. MS1-792US

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TECHNICAL FIELD

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3 This invention generally relates to a technology for deriving robust non-
4 local characteristics and quantizing such characteristics for blind watermarking of
5 a digital good.

6

BACKGROUND

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8 “Digital goods” is a generic label for electronically stored or transmitted
9 content. Examples of digital goods include images, audio clips, video, digital film,
10 multimedia, software, and data. Digital goods may also be called a “digital
11 signal,” “content signal,” “digital bitstream,” “media signal,” “digital object,”
12 “object,” and the like.

13 Digital goods are often distributed to consumers over private and public
14 networks—such as Intranets and the Internet. In addition, these goods are
15 distributed to consumers via fixed computer readable media, such as a compact
16 disc (CD-ROM), digital versatile disc (DVD), soft magnetic diskette, or hard
17 magnetic disk (e.g., a preloaded hard drive).

18 Unfortunately, it is relatively easy for a person to pirate the pristine digital
19 content of a digital good at the expense and harm of the content owners—which
20 includes the content author, publisher, developer, distributor, etc. The content-
21 based industries (e.g., entertainment, music, film, etc.) that produce and distribute
22 content are plagued by lost revenues due to digital piracy.

23 Modern digital pirates effectively rob content owners of their lawful
24 compensation. Unless technology provides a mechanism to protect the rights of
25 content owners, the creative community and culture will be impoverished.

1 **Watermarking**

2 Watermarking is one of the most promising techniques for protecting the
3 content owner's rights of a digital good (i.e., digital good). Generally,
4 watermarking is a process of altering the digital good such that its perceptual
5 characteristics are preserved. More specifically, a "watermark" is a pattern of bits
6 inserted into a digital good that may be used to identify the content owners and/or
7 the protected rights.

8 Watermarks are designed to be completely invisible or, more precisely, to
9 be imperceptible to humans and statistical analysis tools. Ideally, a watermarked
10 signal is perceptually identical to the original signal.

11 A watermark embedder (i.e., encoder) embeds a watermark into a digital
12 good. It typically uses a secret key to embed the watermark. A watermark detector
13 (i.e., decoder) extracts the watermark from the watermarked digital good.

14 **Blind Watermarking**

15 To detect the watermark, some watermarking techniques require access to
16 the original unmarked digital good or to a pristine specimen of the marked digital
17 good. Of course, these techniques are not desirable when the watermark detector
18 is available publicly. If publicly available, then a malicious attacker may get
19 access to the original unmarked digital good or to a pristine specimen of the
20 marked digital good. Consequently, these types of techniques are not used for
21 public detectors.

22 Alternatively, watermarking techniques are "blind." This means that they
23 do not require access to the original unmarked digital good or to a pristine
24 specimen of the marked digital good.

1 specimen of the marked digital good. Of course, these "blind" watermarking
2 techniques are desirable when the watermark detector is publicly available.

3

4 Robustness

5 Before detection, a watermarked signal may undergo many possible
6 changes by users and by the distribution environment. These changes may include
7 unintentional modifications, such as noise and distortions. Moreover, the marked
8 signal is often the subject of malicious attacks particularly aimed at disabling the
9 detection of the watermark.

10 Ideally, a watermarking technique should embed detectible watermarks that
11 resist modifications and attacks as long as they result in signals that are of
12 perceptually the same quality. A watermarking technique that is resistant to
13 modifications and attacks may be called "robust." Aspects of such techniques are
14 called "robust" if they encourage such resistance.

15 Generally speaking, a watermarking system should be robust enough to
16 handle unintentional noise introduction into the signal (such noise may be
17 introduced by A/D and D/A conversions, compressions/decompressions, data
18 corruption during transmission, etc.)

19 Furthermore, a watermarking system should be robust enough and stealthy
20 enough to avoid purposeful and malicious detection, alteration, and/or deletion of
21 the watermark. Such an attack may use a "shotgun" approach where no specific
22 watermark is known or detected (but is assumed to exist) or may use "sharp-
23 shooter" approach where the specific watermark is attacked.

1 This robustness problem has attracted considerable attention. In general, the
2 existing robust watermark techniques fall into two categories: spread-spectrum
3 and quantization index modulation (QIM).

4 With the spread spectrum-type techniques, the watermark indexes the
5 modification to the host data. The host data is the data of the original, unmarked
6 digital signal (i.e., host signal). With typical spread-spectrum watermarking, each
7 bit (e.g., 0s and 1s) of the watermark is embedded into the signal by slightly
8 changing (e.g., adding a pseudorandom sequence that consists of $+\Delta$ or $-\Delta$) the
9 signal.

10 With quantization index modulation (QIM), the watermark is embedded via
11 indexing the modified host data. The modified host data is the data of the marked
12 digital signal (i.e., marked host signal). This is discussed in more detail below.

13 •

14 Those of ordinary skill in the art are familiar with conventional techniques
15 and technology associated with watermarks, watermark embedding, and
16 watermark detecting. In addition, those of ordinary skill in the art are familiar with
17 the typical problems associated with proper watermark detection after a marked
18 signal has undergone changes (e.g., unintentional noise and malicious attacks).

19
20 **Desiderata of Watermarking Technology**

21 Watermarking technology has several highly desirable goals (i.e.,
22 desiderata) to facilitate protection of copyrights of content publishers. Below are
23 listed several of such goals.

1 Perceptual Invisibility. The embedded information should not induce
2 perceptual changes in the signal quality of the resulting watermarked signal. The
3 test of perceptual invisibility is often called the “golden eyes and ears” test.

4 Statistical Invisibility. The embedded information should be quantitatively
5 imperceptible for any exhaustive, heuristic, or probabilistic attempt to detect or
6 remove the watermark. The complexity of successfully launching such attacks
7 should be well beyond the computation power of publicly available computer
8 systems. Herein, statistical invisibility is expressly included within perceptual
9 invisibility.

10 Tamperproofness. An attempt to remove the watermark should damage the
11 value of the digital good well above the hearing threshold.

12 Cost. The system should be inexpensive to license and implement on both
13 programmable and application-specific platforms.

14 Non-disclosure of the Original. The watermarking and detection protocols
15 should be such that the process of proving digital good content copyright both in-
16 situ and in-court, does not involve usage of the original recording.

17 Enforceability and Flexibility. The watermarking technique should provide
18 strong and undeniable copyright proof. Similarly, it should enable a spectrum of
19 protection levels, which correspond to variable digital good presentation and
20 compression standards.

21 Resilience to Common Attacks. Public availability of powerful digital good
22 editing tools imposes that the watermarking and detection process is resilient to
23 attacks spawned from such consoles.

1 **False Alarms & Misses**

2 When developing a watermarking technique, one does not want to increase
3 the probability of a false alarm. That is when a watermark is detected, but none
4 exists. This is something like finding evidence of a crime that did not happen.
5 Someone may be falsely accused of wrongdoing.

6 As the probability of false alarms increases, the confidence in the
7 watermarking technique decreases. For example, people often ignore car alarms
8 because they know that more often than not it is a false alarm rather than an actual
9 car theft.

10 Likewise, one does not want to increase the probability of a miss. This is
11 when the watermark of a signal is not properly detected. This is something like
12 overlooking key piece of evidence at a crime scene. Because of this, a wrongdoing
13 may never be properly investigated. As the probability of misses increases, the
14 confidence in the watermarking technique decreases.

15 Ideally, the probabilities of a false alarm and a miss are zero. In reality, a
16 compromise is often made between them. Typically, a decrease in the probability
17 of one increases the probability of the other. For example, as the probability of
18 false alarm is decreased, the probability of a miss increases.

19 Consequently, a watermarking technique is needed that minimizes both
20 while finding a proper balance between them.

1 **Quantization Index Modulation (QIM)**

2 To that end, some have proposed embedding a watermark by indexing the
3 signal (e.g., host data) during the watermark embedding. This technique is called
4 quantization index modulation (QIM) and it was briefly introduced above.

5 In general, quantization means to limit the possible values of (a magnitude
6 or quantity) to a discrete set of values. Quantization may be thought of as a
7 conversion from non-discrete (e.g., analog or continuous) values to discrete
8 values. Alternatively, it may be a conversion between discrete values with
9 differing scales. Quantization may be accomplished mathematically through
10 rounding or truncation. Typical QIM refers to embedding information by first
11 modulating an index or sequence of indices with the embedded information and
12 then quantizing the host signal with the associated quantizer or sequence of
13 quantizers. A quantizer is class of discontinuous, approximate-identity functions.

14 The major proponent of such QIM techniques is Brian Chen and Gregory
15 W. Wornell (i.e., Chen-Wornell). In their words, they have proposed, “dither
16 modulation in which the embedded information modulates a dither signal and the
17 host signal is quantized with an associated dither quantizer” (from Abstract of
18 Chen-Wornell article from the *IEEE Trans. Inform. Theory*).

19 See the following documents for more details on Chen-Wornell’s proposals
20 and on QIM:

21 • B. Chen and G.W. Wornell, “Digital watermarking and information
22 embedding using dither modulation,” *Proc. IEEE Workshop on*
23 *Multimedia Signal Processing*, Redondo Beach, CA, pp. 273-278,
24 Dec. 1998;

- B. Chen and G. W. Wornell, “Dither modulation: a new approach to digital watermarking and information embedding,” *Proc. of SPIE: Security and Watermarking of Multimedia Contents*, vol. 3657, pp. 342-353, 1999;
- B. Chen and G. W. Wornell, “Quantization Index Modulation: A class of Provably Good Methods for Digital Watermarking and Information Embedding,” *IEEE Trans. Inform. Theory*, 1999 and 2000.

Limitations of Conventional QIM

However, a key problem with conventional QIM is that it is susceptible to attacks and distortions. Conventional QIM relies upon local characteristics within relative to specific representation of a signal (e.g., in the time or frequency domain). To quantize, conventional QIM relies exclusively upon the values of “individual coefficients” of the representation of the signal. An example of such an “individual coefficients” is the color of an individual pixel of an image.

When quantizing, only the local characteristics of an “individual coefficient” are considered. These local characteristics may include value (e.g., color, amplitude) and relative positioning (e.g., positioning in time and/or frequency domains) of an individual bit (e.g., pixel).

Modifications—from either an attack or some type of unintentional noise—can change local characteristics of a signal quite dramatically. For example, these modifications may have a dramatic affect on the color of a pixel or the amplitude of a bit of sound. However, such modifications have little effect on non-local characteristics of a signal.

1 Accordingly, a new and robust watermarking technique is needed to find the
2 proper balance between minimizing the probability of false alarms and the
3 probability of misses, such as QIM watermarking techniques. However, such a
4 technique is needed that is less susceptible to attacks and distortions to the local
5 characteristics within a signal.

6

7 **SUMMARY**

8 Described herein is a technology for deriving robust non-local
9 characteristics and quantizing such characteristics for blind watermarking of a
10 digital good.

11 This technology finds the proper balance between minimizing the
12 probability of false alarms (i.e., detecting a non-existent watermark) and the
13 probability of misses (i.e., failing to detect an existing watermark). One possible
14 technique is quantization index modulation (QIM) watermarking. However,
15 conventional QIM is susceptible to attacks and distortions to the local
16 characteristics of a digital good.

17 The technology, described herein, performs QIM based upon non-local
18 characteristics of the digital good. Non-local characteristics may include statistics
19 (e.g., averages, median) of a group of individual parts (e.g., pixels) of a digital
20 good.

21 This summary itself is not intended to limit the scope of this patent.
22 Moreover, the title of this patent is not intended to limit the scope of this patent.
23 For a better understanding of the present invention, please see the following
24 detailed description and appending claims, taken in conjunction with the
25

1 accompanying drawings. The scope of the present invention is pointed out in the
2 appending claims.

3 **BRIEF DESCRIPTION OF THE DRAWINGS**

4
5 The same numbers are used throughout the drawings to reference like
6 elements and features.

7 Fig. 1 is a schematic block diagram showing a watermarking architecture in
8 accordance with an implementation of the invention claimed herein.

9 Fig. 2 is a schematic block diagram showing an embodiment (e.g., a
10 watermark embedding system) of the invention claimed herein.

11 Fig. 3 is a flow diagram showing an illustrative methodological
12 implementation (e.g., watermark embedding) of the invention claimed herein.

13 Fig. 4 is a schematic block diagram showing an embodiment (e.g., a
14 watermark detecting system) of the invention claimed herein.

15 Fig. 5 is a flow diagram showing an illustrative methodological
16 implementation (e.g., watermark detecting) of the invention claimed herein.

17 Fig. 6 is an example of a computing operating environment capable of
18 implementing an implementation (wholly or partially) of the invention claimed
19 herein.

20 **DETAILED DESCRIPTION**

21
22 In the following description, for purposes of explanation, specific numbers,
23 materials and configurations are set forth in order to provide a thorough
24 understanding of the present invention. However, it will be apparent to one skilled
25 in the art that the present invention may be practiced without the specific

1 exemplary details. In other instances, well-known features are omitted or
2 simplified to clarify the description of the exemplary implementations of present
3 invention, thereby better explain the present invention. Furthermore, for ease of
4 understanding, certain method steps are delineated as separate steps; however,
5 these separately delineated steps should not be construed as necessarily order
6 dependent in their performance.

7 The following description sets forth one or more exemplary
8 implementations of a Derivation and Quantization of Robust Non-Local
9 Characteristics for Blind Watermarking that incorporate elements recited in the
10 appended claims. These implementations are described with specificity in order to
11 meet statutory written description, enablement, and best-mode requirements.
12 However, the description itself is not intended to limit the scope of this patent.

13 The inventors intend these exemplary implementations to be examples. The
14 inventors do not intend these exemplary implementations to limit the scope of the
15 claimed present invention. Rather, the inventors have contemplated that the
16 claimed present invention might also be embodied and implemented in other ways,
17 in conjunction with other present or future technologies.

18 An example of an embodiment of a Derivation and Quantization of Robust
19 Non-Local Characteristics for Blind Watermarking may be referred to as
20 “exemplary non-local QIM watermark.”

1 **Incorporation by Reference**

2 The following co-pending patent applications are incorporated by reference
3 herein (which are all assigned to the Microsoft Corporation):

- 4 • U.S. Patent Application Serial No. 09/390271, entitled "A Technique
5 for Watermarking an Image and a Resulting Watermarked' Image"
6 filed Sept. 7, 1999;
- 7 • U.S. Patent Application Serial No. 09/390272, entitled "A Technique
8 for Detecting a Watermark in a Marked Image" filed on Sept. 7,
9 1999;
- 10 • U.S. Patent Application Serial No. 09/614,660, entitled "Improved
11 Stealthy Audio Watermarking" filed on July 12, 2000,; and
- 12 • U.S. Patent Application Serial No. 09/316,899, entitled "Audio
13 Watermarking with Dual Watermarks" filed on May 22, 1999.

14 **Introduction**

15 The one or more exemplary implementations, described herein, of the
16 present claimed invention may be implemented (in whole or in part) by a non-
17 local QIM watermarking architecture 100 and/or by a computing environment like
18 that shown in Fig. 6.

19 In general, the exemplary non-local QIM watermarker derives robust non-
20 local characteristics of a digital good. It quantizes such characteristics for blind
21 watermarking of the digital good.

22 The exemplary non-local QIM watermarker minimizes the probability of
23 false alarms (i.e., detecting a non-existent watermark) and the probability of
24

1 misses (i.e., failing to detect an existing watermark). It does so by employing
2 quantization index modulation (QIM) watermarking. However, it does not employ
3 conventional QIM techniques because they are susceptible to attacks and
4 distortions to the local characteristics of a digital good.

5

6 Local Characteristics

7 Conventional QIM relies upon local characteristics within a signal (i.e., a
8 digital good). To quantize, conventional QIM relies exclusively upon the values
9 of “individual elements” of the host signal. When quantizing, only the local
10 characteristics of an “individual element” are considered. These local
11 characteristics may include value (e.g., color, amplitude) and relative positioning
12 (e.g., positioning in time and/or frequency domains) of an individual bit (e.g.,
13 pixel).

14 Modifications—from either an attack or unintentional noise—can change
15 local characteristics of a signal quite dramatically. For example, these
16 modifications may have a dramatic affect on the color of a pixel or the amplitude
17 of a bit of sound. However, such modifications have little effect on non-local
18 characteristics of a signal.

19

20 Non-Local Characteristics

21 Non-local characteristics are representative of general characteristics of a
22 group or collection of individual elements. Such a group may be called a segment.
23 Non-local characteristics are not representative of the individual local
24 characteristics of the individual elements; rather, they are representative of the
25 group (e.g., segments) as a whole.

1 The non-local characteristics may be determined by a mathematical or
2 statistical representation of a group. For example, it may be an average of the color
3 values of all pixels in a group. Consequently, such non-local characteristics may
4 also be called “statistical characteristics.” Local characteristics do not have
5 statistical characteristics because a fixed value for a given category. Thus, no
6 statistics are derived from a single value.

7 The non-local characteristics are not local characteristics. They are not
8 global characteristics. Rather, they are in between. Consequently, they may also
9 be called “semi-global” characteristics.

10 **Brief Overview**

11 Given an original, unmarked good, the exemplary non-local QIM
12 watermarker derives robust characteristics that are not local in nature. For
13 example, the exemplary non-local QIM watermarker may employ randomized
14 non-invertible transforms to produce robust non-local characteristics that can be
15 modified without perceptual distortion. These characteristics are typically
16 represented statistically and/or mathematically.

17 To embed the watermark, the exemplary non-local QIM watermarker
18 performs quantization of these non-local characteristics in one or more
19 dimensional lattices or vector spaces. The marked good that results from the
20 exemplary non-local QIM watermarker is robust against unintentional and
21 intentional modifications (e.g., malicious attacks). Examples of malicious attacks
22 include de-synching, random bending, and many other benchmark attacks (e.g.,
23 Stirmark attacks).

1 **Exemplary Non-Local QIM Watermarking Architecture**

2 **Fig. 1** shows a digital goods production and distribution architecture 100
3 (i.e., non-local QIM watermarking architecture 100) having a content
4 producer/provider 122 that produces original content and distributes the content
5 over a network 124 to a client 126. The content producer/provider 122 has a
6 content storage 130 to store digital goods containing original content. The content
7 producer 122 has a watermark embedding system 132 to sign the digital signals
8 with a watermark that uniquely identifies the content as original. The watermark
9 embedding system 132 may be implemented as a standalone process or
10 incorporated into other applications or an operating system.

11 The watermark embedding system 132 applies the watermark to a digital
12 signal from the content storage 130. Typically, the watermark identifies the
13 content producer 122, providing a signature that is embedded in the signal and
14 cannot be cleanly removed.

15 The content producer/provider 122 has a distribution server 134 that
16 distributes the watermarked content over the network 124 (e.g., the Internet). A
17 signal with a watermark embedded therein represents to a recipient that the signal
18 is being distributed in accordance with the copyright authority of the content
19 producer/provider 122. The server 134 may further compress and/or encrypt the
20 content conventional compression and encryption techniques prior to distributing
21 the content over the network 124.

22 Typically, the client 126 is equipped with a processor 140, a memory 142,
23 and one or more content output devices 144 (e.g., display, sound card, speakers,
24 etc.). The processor 140 runs various tools to process the marked signal, such as
25

1 tools to decompress the signal, decrypt the date, filter the content, and/or apply
2 signal controls (tone, volume, etc.). The memory 142 stores an operating system
3 150 (such as a Microsoft® Windows 2000® operating system), which executes on
4 the processor. The client 126 may be embodied in many different ways, including
5 a computer, a handheld entertainment device, a set-top box, a television, an
6 appliance, and so forth.

7 The operating system 150 implements a client-side watermark detecting
8 system 152 to detect watermarks in the digital signal and a content loader 154
9 (e.g., multimedia player, audio player) to facilitate the use of content through the
10 content output device(s) 144. If the watermark is present, the client can identify
11 its copyright and other associated information.

12 The operating system 150 and/or processor 140 may be configured to
13 enforce certain rules imposed by the content producer/provider (or copyright
14 owner). For instance, the operating system and/or processor may be configured to
15 reject fake or copied content that does not possess a valid watermark. In another
16 example, the system could load unverified content with a reduced level of fidelity.

17 **Exemplary Non-Local QIM Watermark Embedding System**

18 Fig. 2 shows an exemplary non-local QIM watermark embedding system
19 200, which is an example of an embodiment of a portion of the non-local QIM
20 watermarking architecture 100. This system may be employed as the watermark
21 encoding system 132 of Fig. 1.

22 The watermark embedding system 200 includes an amplitude normalizer
23 210, a transformer 220, a partitioner 230, segment-statistics calculator 240, a
24 segment quantizer 250, a delta-sequence finder 260, and a signal marker 270.

1 The amplitude normalizer 210 obtains a digital signal 205 (such as an audio
2 clip). It may obtain the signal from nearly any source, such as a storage device or
3 over a network communications link. As its name implies, it normalizes the
4 amplitude of the signal.

5 The transformer 220 receives the amplitude-normalized signal from the
6 normalizer 210. The transformer 220 puts the signal in canonical form using a set
7 of transformations. Specifically, discrete wavelet transformation (DWT) may be
8 employed (particularly, when the input is an image) since it compactly captures
9 significant signal characteristics via time and frequency localization. Other
10 transformations may be used. For instance, shift-invariant and shape-preserving
11 “complex wavelets” and any overcomplete wavelet representation or wavelet
12 packet are good candidates (particularly for images).

13 The transformer 220 also finds the DC subband of the initial transformation
14 of the signal. This DC subband of the transformed signal is passed to the
15 partitioner 230.

16 The partitioner 230 separates the transformed signal into multiple,
17 pseudorandomly sized, pseudorandomly positioned, adjacent, non-contiguous
18 segments (i.e., partitions). A secret key K is the seed for pseudorandom number
19 generation here. This same K may be used to reconstruct the segments by an
20 exemplary non-local QIM watermark detecting system 400.

21 For example, if the signal is an image, it might be partitioned into two-
22 dimensional polygons (e.g., rectangles) of pseudorandom size and location. In
23 another example, if the signal is an audio clip, a two-dimensional representation
24 (using frequency and time) of the audio clip might be separated into two-
25 dimensional polygons (e.g., triangles) of pseudorandom size and location.

1 In this implementation, the segments do not overlap. They are adjacent and
2 non-contiguous. In alternative implementations, the segments may be overlapping.

3 For each segment, the segment-statistics calculator 240 calculates statistics
4 of the multiple segments generated by the partitioner 230. Statistics for each
5 segment are calculated. These statistics may be, for example, any finite order
6 moments of a segment. Examples of such include the mean, the median, and the
7 standard deviation.

8 Generally, the statistics calculations for each segment are independent of
9 the calculations of other segments. However, other alternatives may involve
10 calculations dependent on data from multiple segments.

11 A suitable statistic for such calculation is the mean (e.g., average) of the
12 values of the individual bits of the segment. Other suitable statistics and their
13 robustness are discussed in Venkatesan, Koon, Jakubowski, and Moulin, "Robust
14 image hashing," *Proc. IEEE ICIP 2000*, Vancouver, Canada, September 2000. In
15 this document, no information embedding was considered, but similar statistics
16 were discussed.

17 For each segment, the segment quantizer 250 applies a multi-level (e.g., 2,
18 3, 4) quantization (i.e., high-dimensional, vector-dimensional, or lattice-
19 dimensional quantization) on the output of the segment-statistics calculator 240 to
20 obtain quantized data. Of course, other levels of quantization may be employed.
21 The quantizer 250 may be adaptive or non-adaptive.

22 This quantization may be done randomly also. This may be called
23 randomized quantization (or randomized rounding). This means that the quantizer
24 may randomly decide to round up or round down. It may do it pseudorandomly

1 (using the secret key). This adds an additional degree of robustness and helps hide
2 the watermark.

3 The delta-sequence finder 260 finds a pseudorandom sequence Z that
4 estimates the difference (i.e., delta) between the original transformed signal X and
5 the combination of segments of quantized statistics. The pseudorandom sequence
6 Z may also be called the delta-sequence. For example, if the statistic is averaging,
7 then Z s.t. $\text{Avg}_i(X + Z) = \text{Avg}_i(\hat{X}) = \hat{\mu}_i$, where \hat{X} is the marked signal and $\hat{\mu}_i$ is
8 average for a segment.

9 When finding a delta-sequence Z , it is desirable to minimize the perceptual
10 distortion; therefore, some perceptual distortion metrics may be employed to
11 create this sequence. Thus, in creating Z for, the criteria may include a
12 combination of that minimizes the visual distortions on $X + Z$ (compared to \hat{X})
13 and minimizes the distance between the statistics of $X + Z$ and quantized statistics
14 of X .

15 The signal marker 270 marks the signal with delta-sequence Z so that
16 $\hat{X} = X + Z$. The signal marker may mark the signal using QIM techniques. This
17 marked signal may be publicly distributed to consumers and clients.

18 The functions of aforementioned components of the exemplary non-local
19 QIM watermark embedding system 200 of Fig. 2 are explained in more detail
20 below.

21 **Methodological Implementation of the Exemplary Non-Local QIM**
22 **Watermark Embedding**

23 **Fig. 3** shows the methodological implementation of the exemplary non-
24 local QIM watermark embedding system 200 (or some portion thereof). More
25

1 specifically, this figure shows the methodological implementation of watermark
2 embedding of the exemplary non-local QIM watermarker. This methodological
3 implementation may be performed in software, hardware, or a combination
4 thereof.

5 At 310 of Fig. 3, the exemplary non-local QIM watermarker normalizes the
6 amplitude of the input. The input is the original, unmarked signal (i.e., digital
7 good). At 312, it finds a transform of the amplitude-normalized signal and gets
8 the lowest frequency band (e.g., the DC subband). The result of this block is a
9 transformed signal. This transformation is a discrete wavelet transformation
10 (DWT), but most any other similar transformation may be performed in alternative
11 implementations.

12 At 314, the exemplary non-local QIM watermarker partitions the
13 transformed signal into multiple, pseudorandomly sized, pseudorandomly
14 positioned, adjacent, non-contiguous segments. A secret key K is the seed for
15 pseudorandom number generation here. This same K may be used to reconstruct
16 the segments in the watermark detecting process.

17 For example, if the signal is an image, it might be separated into two-
18 dimensional polygons (e.g., rectangles) of pseudorandom size and location. In
19 another example, if the signal is an audio clip, a two-dimensional representation
20 (using frequency and time) of the audio clip might be separated into two-
21 dimensional polygons (e.g., trapezoids) of pseudorandom size and location.

22 In this implementation, the segments do not overlap. They are adjacent and
23 non-contiguous. In alternative implementations, the segments may be overlapping.

1 For each segment of the signal, the exemplary non-local QIM watermarker
2 repeats blocks 320 through 330. Thus, each segment is processed in the same
3 manner.

4 At 322 of Fig. 3, the exemplary non-local QIM watermarker has a
5 watermark (or a portion thereof) to embed in the current segment. It finds
6 quantized values of statistics of the segment (e.g., averages within the segment or
7 as close as possible). At this point, scalar uniform quantizer and reconstruction
8 points may be randomly perturbed with K as the seed.

9 At 324, the exemplary non-local QIM watermarker finds a pseudorandom
10 sequence Z where the statistics calculation (e.g., averaging) of this sequence
11 combined with the segmented signal generates the calculation of watermarked
12 signal segment. This is also equal to the quantized statistics of the signal segment.

13 At 330, the process loops back to 320 for each unprocessed segment. If all
14 segments have been processed, then it proceeds to 340.

15 At 340 of Fig. 3, the exemplary non-local QIM watermarker combines the
16 pseudorandom sequence with the signal segments to get the watermarked signal
17 segments. In addition, the watermarked signal segments are combined to form a
18 full watermarked signal. In other words, $\hat{X} = X + Z$. At 350, the process ends.

19
20 **Exemplary Non-Local QIM Watermark Detecting System**

21 **Fig. 4** shows an exemplary non-local QIM watermark detecting system
22 400, which is an example of an embodiment of a portion of the non-local QIM
23 watermarking architecture 100. This system may be employed as the watermark
24 detecting system 152 of Fig. 1.

1 The watermark detecting system 400 includes an amplitude normalizer 410,
2 a transformer 420, a partitioner 430, segment-statistics calculator 440, a segment
3 MAP decoder 450, a watermark-presence determiner 460, a presenter 470, and a
4 display 480.

5 The amplitude normalizer 410, the transformer 420, the partitioner 430, and
6 the segment-statistics calculator 440 of the watermark detecting system 400 of Fig.
7 4 function in a similar manner as similarly labeled components of the watermark
8 embedding system 200 of Fig. 2. The exception is that the object of these
9 components is a “subject signal” rather than the original signal. A “subject signal”
10 signal is an unknown. It may or may not include a watermark. It may have been
11 modified.

12 For each segment, the segment MAP decoder 450 determines a quantized
13 value using the MAP decoding scheme. In general, MAP techniques involve
14 finding (and possibly ranking by distance) all objects (in this instance, quantized
15 values) in terms of their distance from a “point.” In this example, the “point” is
16 the statistics calculated for a given segment. Nearest-neighbor decoding is one
17 specific instance of MAP decoding. Those of ordinary skill in the art understand
18 and appreciate MAP and nearest-neighbor techniques.

19 In addition, the segment MAP decoder 450 determines a confidence factor
20 based upon the distance between the quantized values and the statistics calculated
21 for a given segment. If they are coexistent, then factor will indicate a high degree
22 of confidence. If they are distant, then it may indicate a low degree of confidence.
23 Furthermore, the decoder 450 combines the confidence factors of the segments to
24 get an overall confidence factor. That combination may be most any statistical
25 combination (e.g., addition, average, median, standard deviation, etc.).

1 The watermark-presence determiner 460 determines whether a watermark
2 is present. The determiner may use some distortion metric $d(w, \hat{w})$ and some
3 threshold T to decide if watermark is present or not. A normalized Hamming
4 distance may be used.

5 The presenter 470 presents one of three indications: “watermark present,”
6 “watermark not present,” and “unknown.” If the confidence factor is low, it may
7 indicate “unknown.” In addition, it may present the confidence-indication value.

8 This information is presented on the display 480. Of course, this display
9 may be any output device. It may also be a storage device.

10 The functions of aforementioned components of the exemplary non-local
11 QIM watermark detecting system 400 of Fig. 4 are explained in more detail below.

12 Methodological Implementation of the Exemplary Non-Local QIM

13 Watermark Detecting

14 **Fig. 5** shows the methodological implementation of the exemplary non-
15 local QIM watermark detecting system 400 (or some portion thereof). More
16 specifically, this figure shows the methodological implementation of watermark
17 detecting of the exemplary non-local QIM watermarker. This methodological
18 implementation may be performed in software, hardware, or a combination
19 thereof.

20 At 510 of Fig. 5, the exemplary non-local QIM watermarker normalizes the
21 amplitude of the input. The input is an unknown specimen (i.e., signal) of a digital
22 good. It is not known whether this signal includes a watermark or not.

23 At 512, it finds a transform of the amplitude-normalized signal and gets
24 significant frequency subband. That may be a low or the lowest subband (e.g., the

1 DC subband). Generally, the subband selected may be one that represents the
2 signal in a manner that helps further robust watermarking and detection of the
3 watermark. The lower frequency subbands are suitable because they tend to
4 remain relatively invariant after signal perturbation.

5 The result of this block is a transformed signal. When watermarking an
6 image, an example of a suitable transformation is discrete wavelet transformation
7 (DWT). When watermarking an audio clip, an example of a suitable
8 transformation is MCLT (Modulated Complex Lapped Transform). However,
9 most any other similar transformation may be performed in alternative
10 implementations.

11 At 514, the exemplary non-local QIM watermarker partitions the
12 transformed signal into multiple, pseudorandomly sized, pseudorandomly
13 positioned, adjacent, non-contiguous segments. It uses the same secret key K as the
14 seed for its pseudorandom number generation here. Therefore, this process
15 generates the same segments as in the embedding process.

16 For each segment of the signal, the exemplary non-local QIM watermarker
17 repeats blocks 520 through 530. Thus, each segment is processed in the same
18 manner.

19 At 522 of Fig. 5, the exemplary non-local QIM watermarker finds statistics
20 of the segment (e.g., averages within the segment). At 524, the exemplary non-
21 local QIM watermarker finds decoded watermark (or a portion thereof) using
22 Maximum A Posteriori (MAP) decoding. An example of such a MAP decoding is
23 nearest-neighbor decoding.

24 At 526, it measures how close each decoded value is to a quantized value
25 and tracks such measurements. The data provided by such measurements may

1 provide an indication of the confidence of a resulting watermark-presence
2 determination.

3 At 530, the process loops back to 520 for each unprocessed segment. If all
4 segments have been processed, then it proceeds to 540.

5 At 540 of Fig. 5, the exemplary non-local QIM watermarker determines
6 whether a watermark is present. At 542, it determines a confidence-indication,
7 which is based upon the tracked measurement data.

8 At 544, based upon the watermark-presence determination of block 540 and
9 the confidence-indication of block 542, the exemplary non-local QIM watermarker
10 provides one of three indications: watermark present, watermark not present, and
11 unknown. In addition, it may present the confidence-indication value.

12 At 550, the process ends.

13 **Probability of False Alarms and Misses**

14
15 There is a relationship between Probability of false alarm (P_F), Probability
16 of miss (P_M), and the average sizes of the segments. P_F is the probability of
17 declaring that a watermark is present even though it is not. P_M is the probability
18 declaring that a watermark is not present although it is present indeed. Generally,
19 the average segment size is relatively *directly* proportional to the P_F , but it is
20 relatively *indirectly* proportional to the P_M .

21 For example, if the average segment-size is extremely small. So small that
22 they are equivalent to the individual bits (e.g., equal to a pixel in an image. In that
23 situation, watermarks are embedded in single coefficients, which is equivalent to
24 the conventional schemes based on local characteristics. For such a case, P_F is very
25 small and P_M is high. Conversely, if the average segment-size is extremely large.

1 So large that it is the maximum size of the signal (e.g., the whole image). In that
2 situation, P_M is presumably very low whereas P_F is high.

3

4 Other Implementation Details

5 For the following descriptions of an implementation of the exemplary non-
6 local QIM watermark, assume the following:

7 The input signal \underline{X} is an image. Of course, for signal may be other types
8 for other implementations. However, for this example implementation, the signal
9 is an image. Let \underline{w} be the watermark (a binary vector) to be embedded in \underline{X} and
10 a random binary string K be the secret key. The output of a watermark encoder, \hat{X}
11 is the watermarked image in which the information \underline{w} is hidden via the secret key
12 K . This image will possibly undergo various attacks, yielding $\hat{\underline{X}}$. A watermark
13 decoder, given \hat{X} and the secret key K outputs $\hat{\underline{w}}$. It is required that \hat{X} and \underline{X} are
14 approximately the same for all practical purposes and that \hat{X} is still of acceptable
15 quality. The decoder uses some distortion metric $d(\underline{w}, \hat{\underline{w}})$ and some threshold T to
16 decide if watermark is present or not. The exemplary non-local QIM watermark
17 use a normalized Hamming distance, which is the ratio of the usual Hamming
18 distance and the length of the inputs.

19

20 WM via Quantization

21 The exemplary non-local QIM watermark computes a vector $\underline{\mu}$ by
22 applying a forward transformation T_F on \underline{X} . The exemplary non-local QIM
23 watermark assumes that \underline{X} is a grayscale image; if not, a linear transformation
24 may be applied on a colored image to obtain the intensity image. (For colored
25

1 images, most of the energy is concentrated in the intensity plane). Once the
2 information hiding process is done via quantization, there is \hat{u} , and the
3 watermarked data \hat{X} are obtained by applying the transform T_R on \hat{u} . The
4 exemplary non-local QIM watermarker randomizes the process using a random
5 string derived from our secret key K as a seed to a pseudo-random generator.

6 The scheme is generic. So most any transform may be used. In addition,
7 $T_F = (T_R)^{-1}$
8 need not be true, or even that their inverses exist. The exemplary
9 non-local QIM watermarker does not constrain the space where quantization
10 occurs. The decoder applies the transform T_F on the input, \hat{Y} , to get the output is
11 \hat{u}_Y and applying approximate Maximum Likelihood (ML) estimation of the
12 possibly embedded sequence \hat{u}_Y . Through a pseudo-random generator, K will
13 determine many randomization functions of the transform, quantization and
14 estimation stages. Once \hat{u}_Y is found, it is compared with the embedded watermark
15 u , if the distance between them is close to 0 (or less than a threshold T), then it is
16 declared that the watermark is present; otherwise not present. This contrasts with
17 the natural measure for spread spectrum techniques that yields a high value of
18 correlation if the watermark is present.

19 More Details of this Methodological Implementation

20 The transforms T_F and T_R help enhance robustness. For T_F to retain
21 significant image characteristics, we may use discrete wavelet transformation
22 (DWT) at the initial stage. Next, semi-global (i.e., non-local) statistics of segments
23 are determined of the image. The local statistics are not robust.

When the exemplary non-local QIM watermarker computes, for example, the first order statistics of an original image signal and computes several attacked versions for *random* rectangles of fixed size. Then the average mean squared error between these statistics may be found. The error monotonically decays as the size of the rectangles is increased.

In T_F , $\hat{\mu}$ is set to be the estimated first order statistics of randomly chosen rectangles in the wavelet domain. Here T_F is not invertible if the number of rectangles is less than the number of coefficients. To choose T_R , one might first generate a pseudo-random sequence \hat{p} in the image domain that has the property to be visually approximately unnoticeable, pass it through T_F , find the corresponding statistics $\hat{\mu}_p$, compute the necessary scaling factors α such that the pseudorandom sequence \hat{p} scaled by α and added to \hat{X} yields the averages $\hat{\mu}$ that is quantized $\hat{\mu}$. This implementation uses randomly chosen non-overlapping rectangles. The exemplary non-local QIM watermarker defines two quantizers, Q_0 and Q_1 , which map vectors in \mathbb{R}^n to some nearby chosen lattice points.

Here is a formal description. Define $Q_i := \{(q_{ij}, S_{ij}) | j \in \mathbb{Z}\}$, $q_{ij} \in \mathbb{R}^n$ to be *reconstruction points* for Q_i ; $S_{ij} \subset \mathbb{R}^n$ are the corresponding *quantization bins*; $i \in \{0, 1\}$, $j \in \mathbb{Z}$. For each j , then $S_{0j} \cap ((\cup_{k \neq j} S_{0k}) \cup (S_{1k})) = S_{1j} \cap ((\cup_{k \neq j} S_{1k}) \cup (S_{0k})) = \emptyset$. Here q_{0j}, q_{1j} are determined in a pseudo-random fashion (using K as the seed of the random number generator) and n is the dimension of quantization. The exemplary non-local QIM watermarker defines $Q_0[a] = q_{0j} \Leftrightarrow a \in S_{0j}$, and likewise for Q_1 . Both Q_0 and Q_1 are derived from a single quantizer, Q , such that (a) the set of reconstruction points of Q is equal to $\{q_{0j}\}_j \cup \{q_{1j}\}_j$ and (b) for all j, k , $\min_{i \in \mathbb{Z}} d_{L_2}[E(q_{0j}), E(q_{1i})] = \min_{i \in \mathbb{Z}} d_{L_2}[E(q_{0j}), E(q_{0k})]$. Let ξ denote the

1 common value (c) for each fixed k , $\phi_k := \min_{j \neq k} d_{L_2}[E(q_{0j}),$
2 $E(q_{0k})] = \min_{j \neq k} d_{L_2}[E(q_{1j}), E(q_{1k})]$ and $\phi_i = \phi_j > \xi$ for all i, j . Here d_{L_2} is the L_2
3 distortion metric, and $E(\cdot)$ is the expectation operator over the pseudo-random
4 choices q_{ij} .

5 If, for example, $n = 1$. The exemplary non-local QIM watermarker finds
6 $\{E(q_{0j})\}$ and $\{E(q_{1j})\}$ meeting the requirements mentioned above. Then
7 *randomization regions* that are neighborhoods in \mathbb{R}^n around $\{E(q_{0j})\}$ and
8 $\{E(q_{1j})\}$ are introduced. Then $\{q_{0j}\}$ and $\{q_{1j}\}$ are randomly choosen from these
9 regions using some suitable probability distributions. The sizes of the randomizing
10 regions and their shapes are input parameters. Let nR be the number of rectangles
11 each indexed suitably. Let R be the length of the binary watermark vector \underline{w} to be
12 hidden in rectangles. For a vector, \underline{v} denotes its i -th entry by $\underline{v}(i)$. Let L denote
13 the number of levels of DWT that is applied.

14 Encoding

15 Let N be the number of pixels in \underline{X} and $\underline{s} := [s_1, \dots, s_N]$ be the vector that
16 consists of pixels of \underline{X} sorted in ascending order. Create subvectors
17 $\underline{s}_L := [s_t, \dots, s_{t'}]$ where $t := \text{round}(N(1 - \beta)\gamma)$ and $t' = \text{round}(N(1 + \beta)\gamma)$ and
18 $\underline{s}_H := [s_u, s_{u'}]$ where $u = \text{round}(N[1 - (1 - \beta)\gamma])$ and $u' =$
19 $\text{round}(N[(1 + \beta)\gamma])$. Here $\text{round}(r)$ equals the nearest integer to r and the
20 $0 < \beta, \gamma \ll 1$. Let m and M be the mean values of the elements
21 of \underline{s}_L and \underline{s}_H respectively. Apply the point operator
22 $P[x] = 255 * (x - m)/(M - m)$ to each element of \underline{X} to get \underline{X}' .

23
24 Find the L -level DWT of \underline{X}' . Let \underline{X}_A be its DC subband.

1 Partition the \underline{X}_A into random non overlapping nR rectangles; calculate $\underline{\mu}^i$
 2 such that $\underline{\mu}^{(i)}$ is the mean value of the coefficients within rectangle i , $i \leq nR$.
 3 Let $\underline{\mu}^i := [\underline{\mu}(n(i-1)+1), \dots, \underline{\mu}(ni-1), \underline{\mu}(ni)]$ $i \leq R$. Use Q_0 and Q_1 to
 4 quantize: for $i \leq R$, if $\underline{w}(i) = 0$, quantize $\underline{\mu}^i$ using Q_0 else quantize $\underline{\mu}^i$ using Q_1 .
 5 Concatenating the quantized $\{\underline{\mu}^i\}$ we get $\hat{\underline{\mu}}$.

6 Find the differences between the original statistics and the quantized
 7 statistics: $\underline{d} = \hat{\underline{\mu}} - \underline{\mu}$.

8 Generate a pseudo-random sequence $\underline{p} = [p_{ij}]$ in the spatial (i.e. the original
 9 image) domain as follows: Choose p_{ij} randomly and uniformly from $\{0,1\}$ if
 10 $s_{t'} < X_{ij} < s_t$, where $t' = \text{round}(N\gamma)$ and $t = \text{round}(N(1-\gamma))$; otherwise set
 11 $p_{ij}=0$. Now apply L -level DWT to the matrix \underline{p} , extract the DC subband of the
 12 output and call it \underline{p}_W . Now compute the corresponding statistics similar to the
 13 step

14 Compute the scaling factors $\underline{\alpha}$ such that $\underline{\alpha}(i) = \underline{d}(i)/\underline{\mu}_p(i)$, $i \leq nR$.

15 For each rectangle whose index is i , multiply all coefficients of \underline{p}_W within
 16 that rectangle by $\underline{\alpha}(i)$, let the resulting vector be $\hat{\underline{p}}_W$.

17 Apply inverse DWT on $\hat{\underline{p}}_W$, let the output be $\hat{\underline{p}}$.

18 Compute the watermarked data: $\underline{X}'' = \underline{X}' + \hat{\underline{p}}$. Apply the inverse of the
 19 point operator (see step 1) on \underline{X}'' to get $\hat{\underline{X}}$.

20
 21 Decoding (input \underline{Y}):

22 Similar to first part of the encoding above, find the corresponding point
 23 operator on \underline{Y} . Now apply the operator to \underline{Y} and to the output apply L -level
 24 DWT, extract the DC subband and call it \underline{Y}_A .
 25

1 Applying the partitioning procedure of the encoding above to \underline{Y} and find
2 its statistics $\underline{\mu}_Y$. Let $\underline{\mu}_Y^{(i)}$ be the i th element.

3 Using Q_0 and Q_1 , now is described how to carry out approximate ML
4 estimation to find the decoded sequence \underline{w}_Y : Let
5 $\underline{\mu}_Y^i := [\underline{\mu}_Y(n(i-1)+1), \dots, \underline{\mu}_Y(ni-1), \underline{\mu}_Y(ni)]$. For rectangles indexed by
6 $n(i-1)+1, \dots, ni-1, ni, i \leq R$, let $r0(i)$ be the closest point to $\underline{\mu}_Y^i$ among
7 reconstruction points of Q_0 ; likewise let $r1(i)$ be the closest point to $\underline{\mu}_Y^i$ among
8 reconstruction points of Q_1 . If $d_{l2}(\underline{\mu}_Y^i, r0(i)) < d_{l2}(\underline{\mu}_Y^i, r1(i))$ then assign
9 $\underline{w}_Y(i) = 0$; otherwise $\underline{w}_Y(i) = 1$.

10 $d(\underline{w}_Y, \underline{w})$
11 Compute $d(\underline{w}_Y, \underline{w})$. If the result is less than threshold T , declare that the
12 watermark was detected, or otherwise not present.

13 Exemplary Computing System and Environment

14 Fig. 6 illustrates an example of a suitable computing environment 900
15 within which an exemplary non-local QIM watermarker, as described herein, may
16 be implemented (either fully or partially). The computing environment 900 may
17 be utilized in the computer and network architectures described herein.

18 The exemplary computing environment 900 is only one example of a
19 computing environment and is not intended to suggest any limitation as to the
20 scope of use or functionality of the computer and network architectures. Neither
21 should the computing environment 900 be interpreted as having any dependency
22 or requirement relating to any one or combination of components illustrated in the
23 exemplary computing environment 900.

1 The exemplary non-local QIM watermarker may be implemented with
2 numerous other general purpose or special purpose computing system
3 environments or configurations. Examples of well known computing systems,
4 environments, and/or configurations that may be suitable for use include, but are
5 not limited to, personal computers, server computers, thin clients, thick clients,
6 hand-held or laptop devices, multiprocessor systems, microprocessor-based
7 systems, set top boxes, programmable consumer electronics, network PCs,
8 minicomputers, mainframe computers, distributed computing environments that
9 include any of the above systems or devices, and the like.

10 The exemplary non-local QIM watermarker may be described in the
11 general context of computer-executable instructions, such as program modules,
12 being executed by a computer. Generally, program modules include routines,
13 programs, objects, components, data structures, etc. that perform particular tasks
14 or implement particular abstract data types. The exemplary non-local QIM
15 watermarker may also be practiced in distributed computing environments where
16 tasks are performed by remote processing devices that are linked through a
17 communications network. In a distributed computing environment, program
18 modules may be located in both local and remote computer storage media
19 including memory storage devices.

20 The computing environment 900 includes a general-purpose computing
21 device in the form of a computer 902. The components of computer 902 can
22 include, by are not limited to, one or more processors or processing units 904, a
23 system memory 906, and a system bus 908 that couples various system
24 components including the processor 904 to the system memory 906.

1 The system bus 908 represents one or more of any of several types of bus
2 structures, including a memory bus or memory controller, a peripheral bus, an
3 accelerated graphics port, and a processor or local bus using any of a variety of
4 bus architectures. By way of example, such architectures can include an Industry
5 Standard Architecture (ISA) bus, a Micro Channel Architecture (MCA) bus, an
6 Enhanced ISA (EISA) bus, a Video Electronics Standards Association (VESA)
7 local bus, and a Peripheral Component Interconnects (PCI) bus also known as a
8 Mezzanine bus.

9 Computer 902 typically includes a variety of computer readable media.
10 Such media can be any available media that is accessible by computer 902 and
11 includes both volatile and non-volatile media, removable and non-removable
12 media.

13 The system memory 906 includes computer readable media in the form of
14 volatile memory, such as random access memory (RAM) 910, and/or non-volatile
15 memory, such as read only memory (ROM) 912. A basic input/output system
16 (BIOS) 914, containing the basic routines that help to transfer information
17 between elements within computer 902, such as during start-up, is stored in ROM
18 912. RAM 910 typically contains data and/or program modules that are
19 immediately accessible to and/or presently operated on by the processing unit 904.

20 Computer 902 may also include other removable/non-removable,
21 volatile/non-volatile computer storage media. By way of example, Fig. 6
22 illustrates a hard disk drive 916 for reading from and writing to a non-removable,
23 non-volatile magnetic media (not shown), a magnetic disk drive 918 for reading
24 from and writing to a removable, non-volatile magnetic disk 920 (e.g., a “floppy
25 disk”), and an optical disk drive 922 for reading from and/or writing to a

1 removable, non-volatile optical disk 924 such as a CD-ROM, DVD-ROM, or other
2 optical media. The hard disk drive 916, magnetic disk drive 918, and optical disk
3 drive 922 are each connected to the system bus 908 by one or more data media
4 interfaces 926. Alternatively, the hard disk drive 916, magnetic disk drive 918,
5 and optical disk drive 922 can be connected to the system bus 908 by one or more
6 interfaces (not shown).

7 The disk drives and their associated computer-readable media provide non-
8 volatile storage of computer readable instructions, data structures, program
9 modules, and other data for computer 902. Although the example illustrates a hard
10 disk 916, a removable magnetic disk 920, and a removable optical disk 924, it is to
11 be appreciated that other types of computer readable media which can store data
12 that is accessible by a computer, such as magnetic cassettes or other magnetic
13 storage devices, flash memory cards, CD-ROM, digital versatile disks (DVD) or
14 other optical storage, random access memories (RAM), read only memories
15 (ROM), electrically erasable programmable read-only memory (EEPROM), and
16 the like, can also be utilized to implement the exemplary computing system and
17 environment.

18 Any number of program modules can be stored on the hard disk 916,
19 magnetic disk 920, optical disk 924, ROM 912, and/or RAM 910, including by
20 way of example, an operating system 926, one or more application programs 928,
21 other program modules 930, and program data 932. Each of such operating
22 system 926, one or more application programs 928, other program modules 930,
23 and program data 932 (or some combination thereof) may include an embodiment
24 of an amplitude normalizer, a transformer, a partitioner, a segment-statistics
25 calculator, a segment quantizer, an delta-sequence finder, and a signal marker.

1 A user can enter commands and information into computer 902 via input
2 devices such as a keyboard 934 and a pointing device 936 (e.g., a "mouse").
3 Other input devices 938 (not shown specifically) may include a microphone,
4 joystick, game pad, satellite dish, serial port, scanner, and/or the like. These and
5 other input devices are connected to the processing unit 904 via input/output
6 interfaces 940 that are coupled to the system bus 908, but may be connected by
7 other interface and bus structures, such as a parallel port, game port, or a universal
8 serial bus (USB).

9 A monitor 942 or other type of display device can also be connected to the
10 system bus 908 via an interface, such as a video adapter 944. In addition to the
11 monitor 942, other output peripheral devices can include components such as
12 speakers (not shown) and a printer 946 which can be connected to computer 902
13 via the input/output interfaces 940.

14 Computer 902 can operate in a networked environment using logical
15 connections to one or more remote computers, such as a remote computing device
16 948. By way of example, the remote computing device 948 can be a personal
17 computer, portable computer, a server, a router, a network computer, a peer device
18 or other common network node, and the like. The remote computing device 948 is
19 illustrated as a portable computer that can include many or all of the elements and
20 features described herein relative to computer 902.

21 Logical connections between computer 902 and the remote computer 948
22 are depicted as a local area network (LAN) 950 and a general wide area network
23 (WAN) 952. Such networking environments are commonplace in offices,
24 enterprise-wide computer networks, intranets, and the Internet.

1 When implemented in a LAN networking environment, the computer 902 is
2 connected to a local network 950 via a network interface or adapter 954. When
3 implemented in a WAN networking environment, the computer 902 typically
4 includes a modem 956 or other means for establishing communications over the
5 wide network 952. The modem 956, which can be internal or external to computer
6 902, can be connected to the system bus 908 via the input/output interfaces 940 or
7 other appropriate mechanisms. It is to be appreciated that the illustrated network
8 connections are exemplary and that other means of establishing communication
9 link(s) between the computers 902 and 948 can be employed.

10 In a networked environment, such as that illustrated with computing
11 environment 900, program modules depicted relative to the computer 902, or
12 portions thereof, may be stored in a remote memory storage device. By way of
13 example, remote application programs 958 reside on a memory device of remote
14 computer 948. For purposes of illustration, application programs and other
15 executable program components such as the operating system are illustrated herein
16 as discrete blocks, although it is recognized that such programs and components
17 reside at various times in different storage components of the computing device
18 902, and are executed by the data processor(s) of the computer.

19
Computer-Executable Instructions

20 An implementation of an exemplary non-local QIM watermark may be
21 described in the general context of computer-executable instructions, such as
22 program modules, executed by one or more computers or other devices.
23 Generally, program modules include routines, programs, objects, components, data
24 structures, etc. that perform particular tasks or implement particular abstract data
25 structures.

1 types. Typically, the functionality of the program modules may be combined or
2 distributed as desired in various embodiments.

3

4 **Exemplary Operating Environment**

5 Fig. 6 illustrates an example of a suitable operating environment 900 in
6 which an exemplary non-local QIM watermarker may be implemented.
7 Specifically, the exemplary non-local QIM watermarker(s) described herein may
8 be implemented (wholly or in part) by any program modules 928-930 and/or
9 operating system 926 in Fig. 6 or a portion thereof.

10 The operating environment is only an example of a suitable operating
11 environment and is not intended to suggest any limitation as to the scope or use of
12 functionality of the exemplary non-local QIM watermarker(s) described herein.
13 Other well known computing systems, environments, and/or configurations that
14 are suitable for use include, but are not limited to, personal computers (PCs),
15 server computers, hand-held or laptop devices, multiprocessor systems,
16 microprocessor-based systems, programmable consumer electronics, wireless
17 phones and equipments, general- and special-purpose appliances, application-
18 specific integrated circuits (ASICs), network PCs, minicomputers, mainframe
19 computers, distributed computing environments that include any of the above
20 systems or devices, and the like.

21

22 **Computer Readable Media**

23 An implementation of an exemplary non-local QIM watermarker may be
24 stored on or transmitted across some form of computer readable media. Computer
25 readable media can be any available media that can be accessed by a computer.

1 By way of example, and not limitation, computer readable media may comprise
2 “computer storage media” and “communications media.”

3 “Computer storage media” include volatile and non-volatile, removable and
4 non-removable media implemented in any method or technology for storage of
5 information such as computer readable instructions, data structures, program
6 modules, or other data. Computer storage media includes, but is not limited to,
7 RAM, ROM, EEPROM, flash memory or other memory technology, CD-ROM,
8 digital versatile disks (DVD) or other optical storage, magnetic cassettes, magnetic
9 tape, magnetic disk storage or other magnetic storage devices, or any other
10 medium which can be used to store the desired information and which can be
11 accessed by a computer.

12 “Communication media” typically embodies computer readable
13 instructions, data structures, program modules, or other data in a modulated data
14 signal, such as carrier wave or other transport mechanism. Communication media
15 also includes any information delivery media.

16 The term “modulated data signal” means a signal that has one or more of its
17 characteristics set or changed in such a manner as to encode information in the
18 signal. By way of example, and not limitation, communication media includes
19 wired media such as a wired network or direct-wired connection, and wireless
20 media such as acoustic, RF, infrared, and other wireless media. Combinations of
21 any of the above are also included within the scope of computer readable media.

22
23 **Conclusion**

24 Although the invention has been described in language specific to structural
25 features and/or methodological steps, it is to be understood that the invention

1 defined in the appended claims is not necessarily limited to the specific features or
2 steps described. Rather, the specific features and steps are disclosed as preferred
3 forms of implementing the claimed invention.

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